deep-learning-pbl

April 29, 2024

```
[]: %reset -f

[]: from google.colab import drive drive.mount('/content/drive')
```

Mounted at /content/drive

Importing Required Libraries

```
[]: import matplotlib.pyplot as plt
     import pandas as pd
     import os
     import numpy as np
     import tensorflow as tf
     from sklearn.model_selection import train_test_split
     from tensorflow import keras
     import tqdm
     import cv2
     from tensorflow.keras.regularizers import 12
     from tensorflow.keras import layers
     import seaborn as sns
     import gzip
     from tensorflow.keras import layers
     from tensorflow.keras.regularizers import 12
     from tensorflow.keras.preprocessing.image import ImageDataGenerator
     from tensorflow.python.client import device_lib
     from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping, u
      →ReduceLROnPlateau
```

Loading Training/Validation Dataset

```
[]: train_path='/content/drive/MyDrive/seg_train/'
[]: os.listdir(train_path)
```

```
[]: ['buildings', 'forest', 'glacier', 'mountain', 'sea', 'street']
[]: classes=os.listdir(train_path)
[]: length={}
     for n1 in classes:
         l=len(os.listdir(train_path+'/'+n1))
         length[n1]=1
     length
[]: {'buildings': 2191,
      'forest': 2271,
      'glacier': 2404,
      'mountain': 2512,
      'sea': 2274,
      'street': 2382}
[]:|sumi=0
     for e1 in length:
         sumi+=length[e1]
     sumi
[]: 14034
\begin{bmatrix} 1 \\ \end{bmatrix} : \begin{bmatrix} x = \end{bmatrix}
     y=[]
     num_samples_per_class = 1500
     sampled_count_per_class = {labels: 0 for labels in classes}
     for labels in classes:
         pth=os.path.join(train_path,labels)
         for img in tqdm.tqdm(os.listdir(pth)):
            if sampled_count_per_class[labels] >= num_samples_per_class:
                  break
            else:
                pic=cv2.imread(os.path.join(pth,img))
                x.append(pic)
                y.append(classes.index(labels))
            sampled_count_per_class[labels]+=1
                 | 1500/2191 [01:07<00:31, 22.07it/s]
     68% I
     66%|
                 | 1500/2271 [00:55<00:28, 27.05it/s]
                 | 1500/2404 [01:00<00:36, 24.66it/s]
     62%1
                 | 1500/2512 [01:20<00:54, 18.54it/s]
     60%|
                 | 1500/2274 [00:54<00:28, 27.36it/s]
     66%1
     63% l
                 | 1500/2382 [00:59<00:35, 25.09it/s]
```

Pre-Processing of Dataset

```
Resizing to 128x128
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```
[]: for i in range(len(x)):
         img = np.array(x[i])
         x[i] = cv2.resize(img,(128,128))
[]: x = np.array(x)
     y = np.array(y)
[]: np.save('trainX',x)
     np.save('trainY',y)
[]: x.shape
[]: (9000, 128, 128, 3)
[]: y.shape
[]: (9000,)
[]: import numpy as np
     x = np.load("/content/trainX.npy")
     y = np.load("/content/trainY.npy")
    Splitting into training and validation sets
[]: X_train, X_valid, y_train, y_valid = train_test_split(x, y, test_size=0.2,__
      →random_state=42)
[]: X_train= X_train / 255.0
[]: |X_valid = X_valid / 255.0
    Deep CNN Model Initialization
[]: esp = 20
     callback=[
           tf.keras.callbacks.ModelCheckpoint(filepath="/content/drive/MyDrive/CNN/
      ⇔weights/best_model.h5",
      monitor='val_loss', save_best_only=True, verbose=1, save_weights_only=False, mode=|auto'),
           tf.keras.callbacks.EarlyStopping(monitor='val_loss', patience=esp,_
      ⇒restore_best_weights=True),
           tf.keras.callbacks.ReduceLROnPlateau(factor=0.5, patience=5,min_lr=1e-4)
     model = tf.keras.models.Sequential([
         tf.keras.layers.Conv2D(16, (3, 3), activation='relu', input_shape=(128,__
      →128, 3)),
         tf.keras.layers.Conv2D(16, (3, 3), activation='relu'),
```

```
tf.keras.layers.MaxPooling2D((2, 2)),
   tf.keras.layers.Conv2D(32, (3, 3), activation='relu'),
   tf.keras.layers.Conv2D(32, (3, 3), activation='relu'),
   tf.keras.layers.Dropout(0.5),
   tf.keras.layers.MaxPooling2D((2, 2)),
   tf.keras.layers.Conv2D(64, (3, 3), activation='relu'),
   tf.keras.layers.Conv2D(64, (3, 3), activation='relu'),
   tf.keras.layers.Dropout(0.5),
   tf.keras.layers.MaxPooling2D((2, 2)),
   tf.keras.layers.GlobalAveragePooling2D(),
   tf.keras.layers.Dense(32, activation='relu'),
   tf.keras.layers.Dense(32, activation='relu'),
   tf.keras.layers.Dense(6, activation='softmax')
])
model.compile(optimizer='adam', loss='sparse_categorical_crossentropy',
 →metrics=['accuracy'])
```

[]: model.summary()

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 126, 126, 16)	448
conv2d_1 (Conv2D)	(None, 124, 124, 16)	2320
<pre>max_pooling2d (MaxPooling2 D)</pre>	(None, 62, 62, 16)	0
conv2d_2 (Conv2D)	(None, 60, 60, 32)	4640
conv2d_3 (Conv2D)	(None, 58, 58, 32)	9248
dropout (Dropout)	(None, 58, 58, 32)	0
<pre>max_pooling2d_1 (MaxPoolin g2D)</pre>	(None, 29, 29, 32)	0
conv2d_4 (Conv2D)	(None, 27, 27, 64)	18496
conv2d_5 (Conv2D)	(None, 25, 25, 64)	36928
dropout_1 (Dropout)	(None, 25, 25, 64)	0

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max_pooling2d_2 (MaxPoolin (None, 12, 12, 64)
    g2D)
    global average pooling2d (
                         (None, 64)
                                             0
    GlobalAveragePooling2D)
    dense (Dense)
                         (None, 32)
                                             2080
    dense_1 (Dense)
                         (None, 32)
                                             1056
    dense_2 (Dense)
                         (None, 6)
                                             198
   ______
   Total params: 75414 (294.59 KB)
   Trainable params: 75414 (294.59 KB)
   Non-trainable params: 0 (0.00 Byte)
   Model Training & Validation
[]: history = model.fit(X_train,y_train, epochs=150,batch_size=32,__
    →validation_data=(X_valid,y_valid), callbacks=callback)
   Epoch 1/150
   0.4177
   Epoch 1: val_accuracy improved from -inf to 0.49556, saving model to
   /content/drive/MyDrive/CNN/weights/best_model.h5
   /usr/local/lib/python3.10/dist-packages/keras/src/engine/training.py:3103:
   UserWarning: You are saving your model as an HDF5 file via `model.save()`. This
   file format is considered legacy. We recommend using instead the native Keras
   format, e.g. `model.save('my_model.keras')`.
    saving_api.save_model(
   225/225 [========== ] - 19s 44ms/step - loss: 1.3128 -
   accuracy: 0.4183 - val_loss: 1.3129 - val_accuracy: 0.4956 - lr: 0.0010
   Epoch 2/150
   0.5265
   Epoch 2: val_accuracy improved from 0.49556 to 0.51833, saving model to
   /content/drive/MyDrive/CNN/weights/best_model.h5
   accuracy: 0.5272 - val_loss: 1.2472 - val_accuracy: 0.5183 - lr: 0.0010
   Epoch 3/150
```

Epoch 3: val_accuracy improved from 0.51833 to 0.60222, saving model to

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/content/drive/MyDrive/CNN/weights/best_model.h5
225/225 [============= ] - 6s 28ms/step - loss: 0.9860 -
accuracy: 0.5604 - val_loss: 1.1541 - val_accuracy: 0.6022 - lr: 0.0010
Epoch 4/150
0.6042
Epoch 4: val accuracy improved from 0.60222 to 0.62944, saving model to
/content/drive/MyDrive/CNN/weights/best_model.h5
225/225 [============= ] - 6s 26ms/step - loss: 0.9329 -
accuracy: 0.6043 - val_loss: 1.0623 - val_accuracy: 0.6294 - lr: 0.0010
Epoch 5/150
0.6685
Epoch 5: val_accuracy improved from 0.62944 to 0.68000, saving model to
/content/drive/MyDrive/CNN/weights/best_model.h5
225/225 [=========== ] - 6s 29ms/step - loss: 0.8353 -
accuracy: 0.6685 - val_loss: 1.0712 - val_accuracy: 0.6800 - lr: 0.0010
Epoch 6/150
0.6701
Epoch 6: val_accuracy did not improve from 0.68000
accuracy: 0.6703 - val_loss: 1.1256 - val_accuracy: 0.6506 - lr: 0.0010
Epoch 7/150
225/225 [============ ] - ETA: Os - loss: 0.7498 - accuracy:
0.7089
Epoch 7: val_accuracy improved from 0.68000 to 0.73111, saving model to
/content/drive/MyDrive/CNN/weights/best_model.h5
225/225 [============ ] - 7s 30ms/step - loss: 0.7498 -
accuracy: 0.7089 - val_loss: 0.9394 - val_accuracy: 0.7311 - lr: 0.0010
Epoch 8/150
0.7447
Epoch 8: val_accuracy did not improve from 0.73111
accuracy: 0.7453 - val_loss: 0.9464 - val_accuracy: 0.6956 - lr: 0.0010
Epoch 9/150
0.7528
Epoch 9: val_accuracy improved from 0.73111 to 0.76222, saving model to
/content/drive/MyDrive/CNN/weights/best_model.h5
accuracy: 0.7528 - val_loss: 0.8450 - val_accuracy: 0.7622 - lr: 0.0010
Epoch 10/150
225/225 [============= ] - ETA: Os - loss: 0.6222 - accuracy:
Epoch 10: val_accuracy did not improve from 0.76222
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accuracy: 0.7681 - val_loss: 0.9146 - val_accuracy: 0.7400 - lr: 0.0010
Epoch 11/150
Epoch 11: val accuracy improved from 0.76222 to 0.78778, saving model to
/content/drive/MyDrive/CNN/weights/best model.h5
225/225 [============ ] - 7s 31ms/step - loss: 0.5989 -
accuracy: 0.7828 - val_loss: 0.8656 - val_accuracy: 0.7878 - lr: 0.0010
Epoch 12/150
225/225 [============= ] - ETA: Os - loss: 0.5726 - accuracy:
0.7910
Epoch 12: val_accuracy did not improve from 0.78778
accuracy: 0.7910 - val_loss: 0.8749 - val_accuracy: 0.7622 - lr: 0.0010
Epoch 13/150
0.7953
Epoch 13: val_accuracy did not improve from 0.78778
225/225 [============== ] - 6s 28ms/step - loss: 0.5519 -
accuracy: 0.7954 - val_loss: 0.9314 - val_accuracy: 0.7444 - lr: 0.0010
Epoch 14/150
Epoch 14: val_accuracy improved from 0.78778 to 0.79833, saving model to
/content/drive/MyDrive/CNN/weights/best_model.h5
225/225 [============ ] - 7s 30ms/step - loss: 0.5448 -
accuracy: 0.7999 - val_loss: 0.8013 - val_accuracy: 0.7983 - lr: 0.0010
Epoch 15/150
0.8149
Epoch 15: val_accuracy did not improve from 0.79833
accuracy: 0.8149 - val_loss: 0.7725 - val_accuracy: 0.7933 - lr: 0.0010
Epoch 16/150
0.8119
Epoch 16: val accuracy did not improve from 0.79833
accuracy: 0.8118 - val_loss: 0.7751 - val_accuracy: 0.7811 - lr: 0.0010
Epoch 17/150
0.8197
Epoch 17: val_accuracy improved from 0.79833 to 0.80333, saving model to
/content/drive/MyDrive/CNN/weights/best_model.h5
225/225 [=========== ] - 7s 31ms/step - loss: 0.4977 -
accuracy: 0.8197 - val_loss: 0.7727 - val_accuracy: 0.8033 - lr: 0.0010
Epoch 18/150
225/225 [============= ] - ETA: Os - loss: 0.4859 - accuracy:
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0.8242
Epoch 18: val_accuracy did not improve from 0.80333
accuracy: 0.8242 - val_loss: 0.7510 - val_accuracy: 0.7989 - lr: 0.0010
Epoch 19/150
Epoch 19: val_accuracy did not improve from 0.80333
225/225 [============= ] - 6s 27ms/step - loss: 0.4693 -
accuracy: 0.8299 - val_loss: 0.7490 - val_accuracy: 0.7794 - lr: 0.0010
Epoch 20/150
0.8279
Epoch 20: val_accuracy did not improve from 0.80333
225/225 [============ ] - 6s 28ms/step - loss: 0.4768 -
accuracy: 0.8279 - val_loss: 0.7558 - val_accuracy: 0.7878 - lr: 0.0010
Epoch 21/150
0.8383
Epoch 21: val accuracy improved from 0.80333 to 0.81278, saving model to
/content/drive/MyDrive/CNN/weights/best model.h5
accuracy: 0.8383 - val_loss: 0.7314 - val_accuracy: 0.8128 - lr: 0.0010
Epoch 22/150
225/225 [============ ] - ETA: Os - loss: 0.4419 - accuracy:
0.8376
Epoch 22: val_accuracy improved from 0.81278 to 0.81833, saving model to
/content/drive/MyDrive/CNN/weights/best_model.h5
225/225 [=========== ] - 7s 31ms/step - loss: 0.4419 -
accuracy: 0.8376 - val_loss: 0.6967 - val_accuracy: 0.8183 - lr: 0.0010
Epoch 23/150
0.8387
Epoch 23: val_accuracy did not improve from 0.81833
accuracy: 0.8386 - val_loss: 0.7200 - val_accuracy: 0.7917 - lr: 0.0010
Epoch 24/150
225/225 [================= ] - ETA: Os - loss: 0.4184 - accuracy:
0.8489
Epoch 24: val_accuracy did not improve from 0.81833
accuracy: 0.8489 - val_loss: 0.6830 - val_accuracy: 0.8133 - lr: 0.0010
Epoch 25/150
0.8450
Epoch 25: val accuracy improved from 0.81833 to 0.83278, saving model to
/content/drive/MyDrive/CNN/weights/best_model.h5
225/225 [=========== ] - 7s 29ms/step - loss: 0.4223 -
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accuracy: 0.8450 - val_loss: 0.6331 - val_accuracy: 0.8328 - lr: 0.0010
Epoch 26/150
225/225 [============= ] - ETA: Os - loss: 0.4078 - accuracy:
Epoch 26: val accuracy did not improve from 0.83278
accuracy: 0.8501 - val_loss: 0.7041 - val_accuracy: 0.7767 - lr: 0.0010
Epoch 27/150
225/225 [============= ] - ETA: Os - loss: 0.3947 - accuracy:
0.8553
Epoch 27: val_accuracy did not improve from 0.83278
225/225 [============= ] - 6s 27ms/step - loss: 0.3947 -
accuracy: 0.8553 - val_loss: 0.6905 - val_accuracy: 0.7944 - lr: 0.0010
Epoch 28/150
0.8610
Epoch 28: val_accuracy did not improve from 0.83278
225/225 [============ ] - 7s 29ms/step - loss: 0.3878 -
accuracy: 0.8610 - val_loss: 0.6372 - val_accuracy: 0.8089 - lr: 0.0010
Epoch 29/150
0.8569
Epoch 29: val_accuracy did not improve from 0.83278
225/225 [============= ] - 6s 28ms/step - loss: 0.3983 -
accuracy: 0.8568 - val_loss: 0.7077 - val_accuracy: 0.7889 - lr: 0.0010
Epoch 30/150
0.8604
Epoch 30: val_accuracy did not improve from 0.83278
225/225 [============= ] - 6s 28ms/step - loss: 0.3789 -
accuracy: 0.8604 - val_loss: 0.6517 - val_accuracy: 0.8228 - lr: 0.0010
Epoch 31/150
0.8807
Epoch 31: val accuracy did not improve from 0.83278
accuracy: 0.8807 - val_loss: 0.6103 - val_accuracy: 0.8328 - lr: 5.0000e-04
Epoch 32/150
Epoch 32: val_accuracy did not improve from 0.83278
accuracy: 0.8860 - val_loss: 0.6054 - val_accuracy: 0.8189 - lr: 5.0000e-04
Epoch 33/150
225/225 [============= ] - ETA: Os - loss: 0.3104 - accuracy:
Epoch 33: val_accuracy improved from 0.83278 to 0.83444, saving model to
/content/drive/MyDrive/CNN/weights/best_model.h5
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accuracy: 0.8883 - val_loss: 0.5845 - val_accuracy: 0.8344 - lr: 5.0000e-04
Epoch 34/150
0.8901
Epoch 34: val_accuracy improved from 0.83444 to 0.85389, saving model to
/content/drive/MyDrive/CNN/weights/best model.h5
accuracy: 0.8901 - val_loss: 0.5417 - val_accuracy: 0.8539 - lr: 5.0000e-04
Epoch 35/150
225/225 [============= ] - ETA: Os - loss: 0.2989 - accuracy:
Epoch 35: val_accuracy did not improve from 0.85389
accuracy: 0.8947 - val_loss: 0.5965 - val_accuracy: 0.8283 - lr: 5.0000e-04
Epoch 36/150
Epoch 36: val_accuracy did not improve from 0.85389
225/225 [=========== ] - 6s 29ms/step - loss: 0.2953 -
accuracy: 0.8942 - val_loss: 0.5647 - val_accuracy: 0.8344 - lr: 5.0000e-04
Epoch 37/150
Epoch 37: val_accuracy did not improve from 0.85389
accuracy: 0.8964 - val_loss: 0.5337 - val_accuracy: 0.8478 - lr: 5.0000e-04
Epoch 38/150
0.8936
Epoch 38: val_accuracy did not improve from 0.85389
accuracy: 0.8936 - val_loss: 0.5336 - val_accuracy: 0.8406 - lr: 5.0000e-04
Epoch 39/150
0.8976
Epoch 39: val accuracy did not improve from 0.85389
accuracy: 0.8983 - val_loss: 0.5700 - val_accuracy: 0.8356 - lr: 5.0000e-04
Epoch 40/150
0.8975
Epoch 40: val_accuracy did not improve from 0.85389
225/225 [============ ] - 6s 28ms/step - loss: 0.2769 -
accuracy: 0.8975 - val_loss: 0.5150 - val_accuracy: 0.8417 - lr: 5.0000e-04
Epoch 41/150
0.8983
```

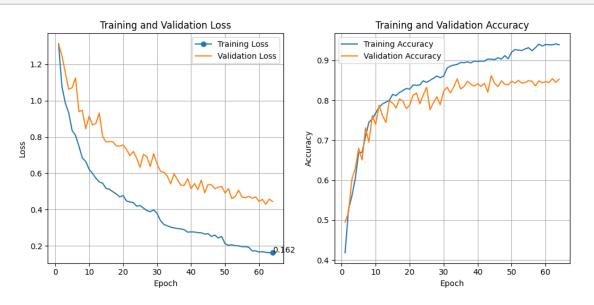
```
Epoch 41: val_accuracy did not improve from 0.85389
accuracy: 0.8983 - val_loss: 0.5429 - val_accuracy: 0.8350 - lr: 5.0000e-04
Epoch 42/150
0.8982
Epoch 42: val accuracy did not improve from 0.85389
accuracy: 0.8982 - val_loss: 0.5097 - val_accuracy: 0.8428 - lr: 5.0000e-04
Epoch 43/150
Epoch 43: val_accuracy did not improve from 0.85389
225/225 [============ ] - 7s 29ms/step - loss: 0.2717 -
accuracy: 0.9033 - val_loss: 0.5623 - val_accuracy: 0.8206 - lr: 5.0000e-04
Epoch 44/150
225/225 [============= ] - ETA: Os - loss: 0.2647 - accuracy:
Epoch 44: val_accuracy improved from 0.85389 to 0.86167, saving model to
/content/drive/MyDrive/CNN/weights/best model.h5
accuracy: 0.9032 - val_loss: 0.4927 - val_accuracy: 0.8617 - lr: 5.0000e-04
Epoch 45/150
0.9019
Epoch 45: val_accuracy did not improve from 0.86167
225/225 [============= ] - 6s 28ms/step - loss: 0.2668 -
accuracy: 0.9019 - val_loss: 0.5379 - val_accuracy: 0.8428 - lr: 5.0000e-04
Epoch 46/150
0.9067
Epoch 46: val_accuracy did not improve from 0.86167
225/225 [============= ] - 6s 28ms/step - loss: 0.2515 -
accuracy: 0.9067 - val_loss: 0.5359 - val_accuracy: 0.8344 - lr: 5.0000e-04
Epoch 47/150
0.9033
Epoch 47: val_accuracy did not improve from 0.86167
accuracy: 0.9033 - val_loss: 0.5146 - val_accuracy: 0.8489 - lr: 5.0000e-04
Epoch 48/150
Epoch 48: val_accuracy did not improve from 0.86167
225/225 [============ ] - 6s 28ms/step - loss: 0.2430 -
accuracy: 0.9119 - val_loss: 0.5230 - val_accuracy: 0.8406 - lr: 5.0000e-04
Epoch 49/150
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0.9043
Epoch 49: val_accuracy did not improve from 0.86167
225/225 [============ ] - 7s 29ms/step - loss: 0.2517 -
accuracy: 0.9043 - val_loss: 0.5273 - val_accuracy: 0.8389 - lr: 5.0000e-04
Epoch 50/150
Epoch 50: val_accuracy did not improve from 0.86167
accuracy: 0.9206 - val_loss: 0.4910 - val_accuracy: 0.8478 - lr: 2.5000e-04
Epoch 51/150
0.9274
Epoch 51: val_accuracy did not improve from 0.86167
225/225 [=========== ] - 6s 28ms/step - loss: 0.2029 -
accuracy: 0.9274 - val_loss: 0.5152 - val_accuracy: 0.8433 - lr: 2.5000e-04
Epoch 52/150
0.9258
Epoch 52: val accuracy did not improve from 0.86167
accuracy: 0.9257 - val_loss: 0.4602 - val_accuracy: 0.8500 - lr: 2.5000e-04
Epoch 53/150
0.9254
Epoch 53: val_accuracy did not improve from 0.86167
225/225 [============ ] - 6s 27ms/step - loss: 0.2004 -
accuracy: 0.9249 - val_loss: 0.4727 - val_accuracy: 0.8433 - lr: 2.5000e-04
Epoch 54/150
0.9290
Epoch 54: val_accuracy did not improve from 0.86167
225/225 [============ ] - 6s 28ms/step - loss: 0.1998 -
accuracy: 0.9287 - val_loss: 0.5063 - val_accuracy: 0.8444 - lr: 2.5000e-04
Epoch 55/150
0.9312
Epoch 55: val_accuracy did not improve from 0.86167
accuracy: 0.9318 - val_loss: 0.4690 - val_accuracy: 0.8494 - lr: 2.5000e-04
Epoch 56/150
Epoch 56: val_accuracy did not improve from 0.86167
225/225 [=========== ] - 7s 29ms/step - loss: 0.1954 -
accuracy: 0.9243 - val_loss: 0.4648 - val_accuracy: 0.8478 - lr: 2.5000e-04
Epoch 57/150
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0.9318
Epoch 57: val_accuracy did not improve from 0.86167
accuracy: 0.9318 - val_loss: 0.4727 - val_accuracy: 0.8361 - lr: 2.5000e-04
Epoch 58/150
Epoch 58: val_accuracy did not improve from 0.86167
225/225 [============ ] - 7s 30ms/step - loss: 0.1718 -
accuracy: 0.9408 - val_loss: 0.4610 - val_accuracy: 0.8489 - lr: 1.2500e-04
Epoch 59/150
0.9361
Epoch 59: val_accuracy did not improve from 0.86167
accuracy: 0.9361 - val_loss: 0.4694 - val_accuracy: 0.8439 - lr: 1.2500e-04
Epoch 60/150
225/225 [============= ] - ETA: Os - loss: 0.1661 - accuracy:
0.9396
Epoch 60: val accuracy did not improve from 0.86167
accuracy: 0.9396 - val_loss: 0.4453 - val_accuracy: 0.8467 - lr: 1.2500e-04
Epoch 61/150
0.9392
Epoch 61: val_accuracy did not improve from 0.86167
accuracy: 0.9392 - val_loss: 0.4558 - val_accuracy: 0.8444 - lr: 1.2500e-04
Epoch 62/150
225/225 [============ ] - ETA: Os - loss: 0.1640 - accuracy:
0.9390
Epoch 62: val_accuracy did not improve from 0.86167
accuracy: 0.9390 - val_loss: 0.4284 - val_accuracy: 0.8544 - lr: 1.2500e-04
Epoch 63/150
Epoch 63: val_accuracy did not improve from 0.86167
accuracy: 0.9415 - val_loss: 0.4574 - val_accuracy: 0.8450 - lr: 1.2500e-04
Epoch 64/150
Epoch 64: val_accuracy did not improve from 0.86167
225/225 [============ ] - 6s 28ms/step - loss: 0.1617 -
accuracy: 0.9393 - val_loss: 0.4438 - val_accuracy: 0.8533 - lr: 1.2500e-04
```

Training & Validation Loss/Accuracy Curves

```
[]: import matplotlib.pyplot as plt
    ep = [i for i in range(1,len(history.history['loss'])+1)]
    curves = [history.history[i] for i in___
      plt.figure(figsize=(10, 5))
    plt.subplot(1, 2, 1)
    plt.plot(ep,curves[0],'o',ls='-',label='Training Loss',markevery=[-1])
    plt.plot(ep,curves[1], label='Validation Loss')
    11 = (ep[-1], curves[0][-1])
    plt.annotate(f"{11[1]:.3f}",xy=11)
    12 = (ep[-1], curves[1][-1])
    plt.title('Training and Validation Loss')
    plt.xlabel('Epoch')
    plt.ylabel('Loss')
    plt.grid()
    plt.legend()
    plt.subplot(1, 2, 2)
    plt.plot(ep,curves[2], label='Training Accuracy')
    plt.plot(ep,curves[3], label='Validation Accuracy')
    plt.title('Training and Validation Accuracy')
    plt.xlabel('Epoch')
    plt.ylabel('Accuracy')
    plt.grid()
    plt.legend()
    plt.tight_layout()
    plt.show()
    # plots for new config model2
```



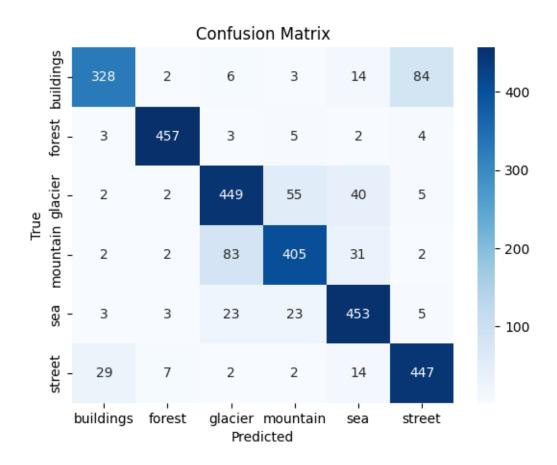
```
Saving Model
 []: bmodel = keras.models.load_model("/content/drive/MyDrive/CNN/weights/
       ⇔best model 0.86167.h5")
     Loading Test Dataset
[27]: test_path='/content/drive/MyDrive/seg_test/seg_test/'
[30]: classes = ['buildings', 'forest', 'glacier', 'mountain', 'sea', 'street']
[31]: lentest={}
      for n1 in classes:
          l=len(os.listdir(test_path+'/'+n1))
          lentest[n1]=1
      lentest
[31]: {'buildings': 437,
       'forest': 474,
       'glacier': 553,
       'mountain': 525,
       'sea': 510,
       'street': 501}
[32]: testimg=0
      for e1 in lentest:
          testimg+=lentest[e1]
      testimg
[32]: 3000
[33]: xtest=[]
      ytest=[]
      for label in classes:
          pth=os.path.join(test_path,label)
          for img in tqdm.tqdm(os.listdir(pth)):
                pic=cv2.imread(os.path.join(pth,img))
                xtest.append(pic)
                ytest.append(classes.index(label))
     100%|
                | 437/437 [00:03<00:00, 131.98it/s]
     100%|
                | 474/474 [00:04<00:00, 107.59it/s]
                | 553/553 [00:04<00:00, 121.69it/s]
     100%|
     100%|
                | 525/525 [00:04<00:00, 123.54it/s]
     100%|
                | 510/510 [00:04<00:00, 112.52it/s]
```

| 501/501 [00:04<00:00, 120.30it/s]

100%|

```
Pre-Processing on Test Dataset
```

```
[34]: for i in range(len(xtest)):
          img = np.array(xtest[i])
         xtest[i] = cv2.resize(img,(128,128))
[35]: xtest = np.array(xtest)/255
      ytest = np.array(ytest)
      np.save('testX',xtest)
      np.save('testY',ytest)
[36]: import numpy as np
      xtest = np.load("/content/testX.npy")
      ytest = np.load("/content/testY.npy")
[37]: ytest
[37]: array([0, 0, 0, ..., 5, 5, 5])
     Results of Model Testing
[38]: ypred = bmodel.predict(xtest,batch_size=32)
     94/94 [========] - 1s 7ms/step
[39]: ypredi = np.argmax(ypred,axis=1)
      ypredi
[39]: array([0, 0, 0, ..., 5, 5, 5])
[40]: from sklearn.metrics import
      Goonfusion_matrix,classification_report,accuracy_score,f1_score
      cm = confusion_matrix(ytest,ypredi)
      sns.heatmap(cm, annot=True, fmt="d", cmap='Blues', xticklabels=classes, __
       ⇔yticklabels=classes)
      plt.xlabel("Predicted")
      plt.ylabel("True")
      plt.title("Confusion Matrix")
      plt.show()
```



```
[41]: acc = accuracy_score(ytest,ypredi)
    print(f"Accuracy = {acc*100:.2f}%")

Accuracy = 84.63%

[42]: f1 = f1_score(ytest,ypredi,average='macro')
    print(f"F1-Score = {f1*100:.2f}%")

F1-Score = 84.73%

[43]: rep = classification_report(ytest,ypredi,target_names=classes)
    print(rep)
```

	precision	recall	f1-score	support
buildings	0.89	0.75	0.82	437
forest	0.97	0.96	0.97	474
glacier	0.79	0.81	0.80	553
mountain	0.82	0.77	0.80	525
sea	0.82	0.89	0.85	510

street	0.82	0.89	0.85	501
accuracy	0.05	0.05	0.85	3000
macro avg	0.85	0.85	0.85	3000
weighted avg	0.85	0.85	0.85	3000

[]: